Optimization notes

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Directional derivative

From a starting point \underline{x}_0 and a given direction \underline{u} :

- $\underline{x}(\lambda) = \underline{x}_0 + \lambda \underline{u}$
 - λ is a scalar.
- $d\underline{x} = \underline{u}d\lambda$
 - For a small change in λ .
- $F(\lambda) = f(\underline{x}_0 + \lambda \underline{u})$

$$dF = df = (\nabla f(\underline{x}))^{\top} d\underline{x}$$
$$= (\nabla f(\underline{x}))^{\top} \underline{u} d\lambda = \nabla^{\top} f \underline{u} \lambda$$

- $\frac{df}{d\lambda} = \nabla^{\top} f \underline{u}$
 - If f is minimized at $\underline{x}^* = \underline{x}_0 + \lambda \underline{u}$, then:
 - $\nabla f(\underline{x}^*))^{\top} f\underline{u} = 0$
 - gradient f evaluated at the minimum point is orthogonalto \underline{u} .

Weierstrass Theorem

If $f(\underline{x})$ is continuous on a nonempty feasible set that is cloased and bounded, then $f(\underline{x})$ has a global minimum in this set.

- ▶ A set *S* is bounded if for any point \underline{x} in *S*, we have $\underline{x}^{\top}\underline{x} < c$
 - c is a finite positive number.

Single-variable unconstrained optimization

- Necessary condition
 - If a function f(x) has a local minimum at $x = x^*$, and f'(x) exists as a finite number at $x = x^*$, then $f'(x^*) = 0$.
 - x^* at $f'(x^*) = 0$ is called stationary point.
- Sufficient condition
 - Suppose $f'(x^*) = f''(x^*) = \cdots = f^{(m-1)}(x^*) = 0$, but $f^{(m-1)}(x^*) \neq 0$, then $f(x^*)$ is:
 - 1. a local minimum if $f^{(m-1)}(x^*) > 0$ and m is even.
 - 2. a local maximum if $f^{(m-1)}(x^*) < 0$ and m is even.
 - 3. neither a maximum nor a minimum if *m* is odd.

Multi-variable unconstrained optimization (1)

Definition of r^{th} differential of function f:

$$d^{r}f(\underline{x}^{*}) = \sum_{i=1}^{n} \sum_{j=1}^{n} \cdots \sum_{k=1}^{n} h_{i}h_{j} \dots h_{k} \frac{\partial^{r}f(\underline{x}^{*})}{\partial x_{i}\partial x_{j} \dots \partial x_{k}}$$

Example

When (order) r = 2 and (number of variables) n = 3, we have:

$$d^{2}f(\underline{x}^{*}) = d^{2}f(x_{1}^{*}, x_{2}^{*}, x_{3}^{*}) = \sum_{i=1}^{3} \sum_{j=1}^{3} h_{i}h_{j} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{i}\partial x_{j}}$$

$$= h_{1}^{2} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{1}^{2}} + h_{2}^{2} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{2}^{2}} + h_{3}^{2} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{3}^{2}}$$

$$+ 2h_{1}h_{2} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{1}\partial x_{2}} + 2h_{2}h_{3} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{2}\partial x_{3}} + 2h_{1}h_{3} \frac{\partial^{2}f(\underline{x}^{*})}{\partial x_{1}\partial x_{3}}$$

Multi-variable unconstrained optimization (2)

Necessary condition

$$\frac{\partial f(\underline{x}^*)}{\partial x_1} = \frac{\partial f(\underline{x}^*)}{\partial x_2} = \dots = \frac{\partial f(\underline{x}^*)}{\partial x_n} = 0$$

- In vector form, $\nabla f(\underline{x}^* = 0)$.
- \underline{x}^* at $\nabla f(\underline{x}^* = 0)$ is called stationary point.
- Sufficient condition
 - For a stationary point at $\underline{x} = \underline{x}^*$:
 - if the Hessian matrix of $f(\underline{x})$ evaluated at $\underline{x} = \underline{x}^*$ is positive definite, then x^* is a local minimum.
 - if the Hessian matrix of $f(\underline{x})$ evaluated at $\underline{x} = \underline{x}^*$ is negative definite, then \underline{x}^* is a local maximum.